

Itemized Response

Reg. To: US Patent Application # 10/662,552

Zhang, Xiaohui, 4400 E Broadway blv, Tucson, AZ 85711 Fax: 520-202-3340; Phone: 520-202-3333 Applicant:

Methods And System For Bio-Intelligence From Over-The-Counter Pharmaceutical Sales

Mail Date: March 16, 2007

Examiner: Mr. Neal Sereboff, Art Unit 3626

Note of Response:

Reference U is Goldenberg et al, Early statistical detection of anthrax outbreaks by tracking over-the-counter medication sales, Proc Natl Acad Sci U S A. 2002 April 16;

Reference V is Armstrong et al, Updated Guidelines for Evaluating Public Health Surveillance Systems, CDC July 27, 2001 / 50(RR13);1-35

Reference [3] is the attached reference by X. Zhang et all, A Biointelligence System for Identifying Potential Disease Outbreaks, IEEE Engineering in Medicine and Biology (EMB), Vol 23, Number 1, 2004.

Response

| - | The second secon | |
|----------|--|--|
| Action # | Action # Requirement/Question / Advise by The Examiner | Analysis & Response from the Applicant |
| 1 | Claim 1 -15 are pending | Response within 3 months |
| 2 | Advise to have a patent attorney | Applicant will have a patent attorney soon. |
| 3 | Claim 6 and Claim 9 in improper form because | Corrected Claim 6 and Claim 9 to the alternative form, following |
| | of multiple dependent claim should be in the | the examples in Section 7.45 of MPEP 608.01(n) |
| | alternative | |
| 4 | Appropriate correction is required for verb | Corrected verb 'are' to 'is' |
| 5 | Claim 11 partially repeated the language of | Removed the repeated sentence. |
| | Claim 10, Appropriate correction is required | |
| | Carrier 19, 19 Probremo correction is reducing | , |

| Action # | Requirement / Question / Advise by The | Analysis & Response from the Applicant |
|----------|---|---|
| | Examiner | |
| 9 | Amend the claim, since Claim 15 dependent | Removed the sentence that is dependent on Claim 12 |
| | upon Claim 1, not Claim 12 | |
| 7 | Copy of '35 U.S.C. 101 | Noted |
| 8 | 'non-statutory subject matter' | The claims 1-15 represent a new useful system, its application |
| | | demonstrated that it is statutory subject matter according to 35 U.S.C. 101. For elaboration please review the response letter, the |
| | | attached reference [3] and responses below. |
| 6 | (1) Question of 'the results be | (1) The system and the method in claim 1-15 have been |
| | reproducible | implemented in a computer system, and said system has |
| | | been helping public health workers in identifying disease |
| | | outbreaks. The results are completely reproducible for the |
| | | given data and conditions. Please see the attached reference |
| | | publication by X. Zhang et al: A Biointelligence System for |
| | | Identifying Potential Disease Outoreaks, in IEEE |
| | | Number 1, 2004. |
| | (2) Question on 'Subjective component' | |
| | | (2) There is no 'subjective component' in this system; it intends |
| | | to overcome the limits from 'subjective components'. The |
| | | results are completely based on an analysis of the data and |
| | | are not dependent on subjective feelings of the analyst. The |
| | | rules are derived directly from the investigated data itself |
| | | not from "experience", or other subjective sources. The |
| | | system automatically takes the historical data to derive the |
| | | categorized public health status, reference values, decide the |
| | | state transitions, and make decisions, in a specified place. |
| 10 | Implementation of the rule system | (1) The system has been implemented in a computer system, |
| | | |

| Analysis & Response from the Applicant | including the rule systems. Several of the included figures are taken from sample implementations. Please see the attached reference publication by X. Zhang et al: A Biointelligence System for Identifying Potential Disease Outbreaks, in IEEE Engineering in Medicine and Biology (EMB), Vol 23, Number 1, 2004. | (2) In the patent application, the rule systems are demonstrated in detail (please see paragraph 86 to paragraph 93), step by step, the rule system decides the state transition and inputs/outputs with data referenced in fig. (5) to (8). In fact, the same examples were provided to two different software developers, and they both implemented their application systems independently, one system was published by X. Zhang et al: A Biointelligence System for Identifying Potential Disease Outbreaks, in IEEE Engineering in Medicine and Biology (EMB), Vol 23, Number 1, 2004. The other system is currently in | Noted | |
|---|--|--|--|---|
| Requirement / Question / Advise by The Examiner | | | Copy of first paragraph of 35 U.S.C. 112 | "The specification shall contain a written description of the invention, and of the manner and process of making and using it, in such full, clear, concise, and exact terms as to enable any person skilled in the art to which it pertains, or with which it is most nearly connected, to make and use the same, and shall set forth the best mode contemplated by the inventor of carrying out his invention." |
| Action # | · | | 11 | |

| Action # | nent // Question // Advise by The | Analysis & Response from the Applicant |
|----------|--|---|
| | Question (1) how the invention overcomes existing problems (Background of Invention) | (1) Existing system consider a categorized public health status in a community (such as flu situation) as one of two status: 'normal' or 'outbreak'. Traditionally, analyses are |
| | | conducted in such binary mode. This is a shortcoming that results from the use of traditional statistical approaches. The usual objective is to try to detect an 'outbreak' (one out of the two status) early. However, the system presented in |
| | | claim 1-15 we model for the first time a dynamic change of a categorized public health status in a community by a set of state variables (seven state variables) and the transition of |
| | | state variables. The seven state variables defined by the applicant are the minimum set to characterize a complete process of a community's public health status. This |
| | | approach is inherently different from statistical approaches as it integrates dynamic model, rule systems and information technique. |
| | (2) lack of method of practicing and working example | (2) The presented system has been described in sufficient detail as a blue print for software engineers to implement as evident in the two independent and successful implementations, which |
| | | both include the outlined rule systems. Please see the attached reference publication by X. Zhang et al: A Biointelligence System for Identifying Potential Disease Outbreaks, in IEEE Engineering in Medicine and Biology (EMB), Vol 23, Number 1, 2004. |

| Action # | Requirement / Question / Advise by The Examiner | Analysis & Response from the Applicant |
|----------|---|--|
| 13 | 'claim 1 – 15 are rejected, based on a disclosure is not enabling' | The system is implemented and already deployed; the implemented system has been running since 2003 and has successfully aided a Public Health department in the cognition of several otherwise unknown disease outbreak events. Please see the attached reference publication by X. Zhang et al: A Biointelligence System for Identifying Potential Disease Outbreaks, in IEEE Engineering in Medicine and Biology (EMB), Vol 23, Number 1, 2004. |
| | 'structural implementation' | Examples of implementation are provided in the attached publication by X. Zhang et al: A Biointelligence System for Identifying Potential Disease Outbreaks, in IEEE Engineering in Medicine and Biology (EMB), Vol 23, Number 1, 2004. |
| 14 | Claim 1-15 are rejected, as failing to comply with enablement requirement. working example or instruction | Working example is provided in the attached publication by X. Zhang et al: A Biointelligence System for Identifying Potential Disease Outbreaks, in IEEE Engineering in Medicine and Biology (EMB), Vol 23, Number 1, 2004. |
| 15 | quotation of the second paragraph of 35 U.S.C. 112: "The specification shall conclude with one or more claims particularly pointing out and distinctly claiming the subject matter which the applicant regards as his invention." | Noted Please see response to Action 16. |
| | | |

| Action # | Requirement / Onestion / Advise by The | Anslysis & Resnonse from the Annlicent |
|----------|--|--|
| | Examiner | The state of the s |
| 16 | Need to define the invention required by 15 | Two of obvious inventions are identified as examples: |
| | | It is the first time (an invention), that a dynamic change of a categorized public health status in a community is modeled by a set |
| · | | or state variables (seven variables) and the transition of state variable in a dynamic model. The seven state variables defined by the applicant are the minimum set to characterize a complete |
| | | process of a community's categorized public health status. |
| | | It is the first time, which in dynamic modeling in state space approach, the state transition matrix is systematically defined by a |
| | | combination of numerical functions and rule base and knowledge management. |
| | | |
| 17 | Language and format | |
| | The claim must be in one sentence form only | Modified format. |
| 18. | Use of 'the' and 'a' in Claim 1 | Corrected 'the rule system' to 'a rule system' in Claim 1 |
| 19 | Use of 'the' and 'a' in Claim 1 | Corrected |
| 20 | Use of 'the apparatus' | Corrected in Claim 1 |
| 21 | Use of 'the' baseline | Corrected to 'a' |
| 22 | Claim 3 -5 failing to point out and distinctly | |
| | claim the invention | |
| | Claim 3 -5 dependent upon Claim 2 | Modified Claim 3 – 5. |
| 23 | Claim 10 failing to point out and distinctly claim the invention | Mended the sentence, as |
| | | |

| Action # | Action # Requirement/ Question / Advise by The Examiner | Analysis & Response from the Applicant |
|----------|--|---|
| · | Question: what is being mapped | The structural components are mapped incorporating their confidence levels. |
| 24 | quotation of the paragraph of 35 U.S.C. 102: "A person shall be entitled to a patent unless – | Noted. |
| | (b) the invention was patented or described in a printed publication in this or a foreign country or in public use or on sale in this country, more than one year prior to the date of the application for patent in the United States." | |
| 25 | Claim 1 – 5, 7, 10-15 are rejected, because of Goldenberg et al "Farly statistical detection of | Please see the response to Action 26. |
| | anthrax outbreaks by tracking over-the-counter medication sales", reference U | |
| 26 | As per claim 1, | In reference U, Goldenberg, as all other public health workers, does |
| | The first paragraph: | not define public health status as a set of state variables, does not model 'the change of categorized public health status', there is no word of 'nublic health status' appeared in Reference 11 |
| | 'Goldenberg teaches the system for detecting an unusual public health status and for modeling the change of categorized public | Goldenberg tries to use OTC data to detect an outbreak only |
| | health status from OTC pharmaceutical data' | to model the detection in the space of state variables as described in Claim 1. |
| | | Goldenberg's approach is the traditional statistical approach with a signal detection for Normal or Attack (outbreak); while the system presented in claim 1-15 describes a complete process for a categorized public health status in a community. |
| | | |

| Requirement / Question / Advise by The Analysis & Response from the Applicant Examiner |
|---|
| Claim 1 is different from the one in the reference. |
| Quotation of Tracking Grocery Data section, paragraph 1 (of reference U): |
| "Tracking Grocery Data Grocery and OTC medication sales have three main advantages for the detection of an outbreak: First, these datasets are typically very large and rich, including information on each purchased item and in many cases include customer information(e.g., address). They are also available on a more frequent scale, such as daily and even hourly basis, and do not include delays in reporting as compared with medical and public health sources which are typically collected weekly or even less frequently, and might contain delays. Second, the outbreak footprint would probably exist in these data earlier than in medical or public health data, because of self treatment that people usually pursue before seeking medical assistance. Third, although grocery and OTC sales do not measure illness directly, we might infer specific symptoms experienced by purchasers at a relatively early stage of the onset of the disease." |
| Analysis: In this paragraph, Goldenberg summarized the advantages of using OTC data. We agree with his analysis of the benefits, which is why we are working with the same data source. No specific method is presented. Goldenberg mentioned the use of customer information (e.g. address). Immediately in the following paragraph, Goldenberg pointed out 'the main problem using grocery and OTC medication sales is their noise nature'. |
| In Claim 1 – 15, no a customer address is required or used. In section of Summary of Invention, a study area or a geographical level is defined by a pharmacy store service area, a zip code area, a city, a county, and a statewide area; while the application of the claimed system and method is the same. |

| Action # | Requirement / Question / Advise by The Examiner | Analysis & Response from the Applicant |
|----------|--|---|
| | | The system of variables and calculations demonstrated elsewhere in Goldenberg is of a mathematically very different nature with all the discussed shortcomings of statistics based approaches to the "noise" in data. |
| | | Conclusion: |
| | | The claims are different from the one in the reference. |
| | The third paragraph | Quotation of reference U, page 2, Tracking Grocery Data, |
| | An algorithm for unusual public health status (or event) detection incorporating seasonally varying reference lines (see page 2, Tracking Grocery Data, paragraph 5), and calculating three structural components from | paragraph 5: "Our proposed detection system consists of several layers (A.G., G.S., and R.A.C., unpublished results). The first layer preprocesses the data by accounting for store level sales. The second layer puts the preprocessed data through a denoising filter. We use the discrete cosine transform (10, 11), which decomposes the series into cosine waves, and our filter retains only those that have a large magnitude. We chose the number of retained cosine waves to capture the main features of the series but also to avoid overfitting (A.G., G.S., and R.A.C., unpublished results)." |
| | the input data: a daily deviation from the reference line (see Figure 3), an n-days-cumulated-deviation, and | Analysis: In reference U, (page 2, Tracking Grocery Data, paragraph 5) Goldenberg uses their 'unpublished results', and a wavelet transform. which is totally different from the approach in the Claim |
| | (see Figure 3) • the change of the daily deviations in | 1. |
| | that area, (see Figure 3) | In reference U, Figure 3, Goldenberg compares the sales data with the threshold which 'is in fact three standard deviations', as |
| | | Goldenberg writes in Page 3, the second paragraph (above Figure |
| | | 3), as a typical statistic approach. Goldenberg does not mention an |
| | | n-days-cumulated-deviation in Figure 3, not in his paper either; Goldenberg does not mention the change of the daily deviation in |

| Action # | Requirement/Question/Advise by The Examiner | Analysis & Response from the Applicant |
|-------------|--|--|
| | | Figure 3, not his paper either. |
| | | Conclusion: the claims are different from the one in the reference. |
| | The fourth paragraph | Copy of Page 3, paragraph 1 of Reference U: |
| | A dynamic system model describing the categorized public health status by a set of | rust, we decompose are denoised series into several resolutions by using a discrete (redundant) wavelet transform (ref. 12; cf. the continuous version of wavelets in ref. 13). Each resolution describes a different frequency of the series, but, unlike other transforms (e.g., the cosine and Fourier transform), it retains information on the <i>time</i> that each |
| | status by the state transition, the input sets, the output sets, and the rule systems that | use a simple autoregressive model (where the sales at time t are taken to be a weighted average of previous sales) for predicting each resolution separately. We then add the prediction to create the forester of the cast and the creative to create the forester of the cast and the creative to create the forester of the cast and the creative to create the forester of the cast and the creative to create the forester of the cast and the creative to create the forester of the cast and the creative to create the forester of the cast and the creative transfer of the cast and the creative transfer of the cast and the creative transfer of the cast and the c |
| - | govern them (see Page 3 paragraph 1) | of the (preprocessed and denoised) series into five resolutions. For each resolution, we use an autoregressive model for forecasting the next point. Finally, we add the forecasts to obtain the next point in the series, i.e., Fig. 2 also includes the combined forecast of the |
| | | next day (denoised) sales." |
| · · · · · · | | Analysis: In Page 3, paragraph 1 of Reference U, Goldenberg describes 'a |
| | | wavelet transform' and 'a simple autoregressive model'. His transform is a mathematical treatment/model of the problem which |
| | | has no physical meaning nor describes a public health concept or content; his autoregressive model is for 'forecast of the next day |
| · | | sales'. The proposed patent matches each state variable with a |
| | | public health concept/content, which is very unique to the proposed |
| | | parent and possion with nothing inducting approaches. |
| | | Goldenberg does not mention a model with a set of state variables, |
| | | and does not model a change of the public health status by the state transition either. |

| | the reference. | | by using a discrete avelets in ref. 13). The other transforms are that each regular, and thus we to be a weighted e then add the e decomposition h resolution, we use I the forecasts to ed forecast of the | lescribes 'a lel'. His problem which | h concept or he next day | ble with a ressive model | | d the method he | onal statistical | ule aynamic |
|---|---|---|--|--|---|---|--|---|--|--|
| Analysis & Response from the Applicant | Conclusion: the claims are different from the one in the reference. | Copy of Page 3, paragraph 1 of Reference U: | "First, we decompose the denoised series into several "resolutions" by using a discrete (redundant) wavelet transform (ref. 12; cf. the continuous version of wavelets in ref. 13). Each resolution describes a different frequency of the series, but, unlike other transforms (e.g., the cosine and Fourier transform), it retains information on the <i>time</i> that each frequency is present. The resulting series for each resolution are more regular, and thus we use a simple autoregressive model (where the sales at time t are taken to be a weighted average of previous sales) for predicting each resolution separately. We then add the predictions to create the forecast of the next day sales. Fig. 2 shows the decomposition of the (preprocessed and denoised) series into five resolutions. For each resolution, we use an autoregressive model for forecasting the next point. Finally, we add the forecasts to obtain the next point in the series, i.e., Fig. 2 also includes the combined forecast of the next day (denoised) sales." | Analysis: In Page 3, paragraph 1 of Reference U, Goldenberg describes 'a wavelet transform' and 'a simple autoregressive model'. His | has no physical meaning nor describes a public health concept or content; his autoregressive model is for 'forecast of the next day | sales'. The proposed patent matches each state variable with a public health concept/content. Goldenberg's autoregressive model | attempts to create the 'forecast of the next day sales'. | Goldenberg does not mention the state transition, and the method he | used here (simple autoregressive model) is a traditional statistical | change of public health status either. |
| Analysis & Respo | Conclusion: the cl | Copy of Page 3, par | "First, we decompose the (redundant) wavelet transfor Each resolution describes a (e.g., the cosine and Fourie frequency is present. The ruse a simple autoregressive average of previous sales) to predictions to create the for of the (preprocessed and de an autoregressive model for obtain the next point in the next day (denoised) sales." | Analysis: In Page 3, paragra wavelet transform transform is a mat | has no physical me | sales'. The propos public health conc | attempts to create | Goldenberg does 1 | used here (simple | change of public h |
| Requirement / Question / Advise by The Examiner | | The fifth paragraph | The rule system determines the state transitions for modeling the dynamic change of public health status through the analysis of information derived from OTC pharmaceutical sales in that area (see page 3, paragraph 1) | | | | | | | |
| Action # | · | | | | | - | | | | |

| v | | | | |
|--|---|--|--|--|
| Analysis & Response from the Applicant | Conclusion: the claims are different from the one in the reference. | Quotation of paragraph 2 in page 3 of reference U: "The final layer of the detection system includes the computation of an upper threshold for the next day forecasts. This threshold is based on the forecast made in the previous step, plus a margin of error. When the actual next day sales become available, they are compared with the threshold. If they exceed the threshold, the system flags an alarm, indicating that the new daily sales are higher than expected. The threshold is based on the distribution of the differences between the forecasts and the real sales, and is in fact three standard deviations of the differences above the denoised series. This last step is based on a methodology used in statistical quality control, called control charts, where a process is monitored by using a chart that flags when a change occurs, while taking into account natural variation of the series (14). Fig. 3 illustrates the threshold for the cough OTC medication data. The threshold follows the series, creating a "security band," which, if exceeded, is an indication that the sales are higher than expected. For example, sales for 8_7_00 are higher than the prediction. They do not exceedthe threshold, however, and thus we do not take them to indicate an abnormal increase in sales." | Analysis: In the Claim 1, three structured components are defined. The claim further describes a rule system combining the three components with confidence supporting sets (where the confidence supporting sets are defined by equation 7 - 9 in later section) as the input variables. The input variables are different from input data which is raw sale data. Also note, that confidence supporting set in the proposed method is not equivalent to 'three standard deviation'. | In reference U, Goldenberg does not have those three components, no confidence supporting sets of those three components either. In reference U, page 3, paragraph 2, Goldenberg uses a threshold, |
| Requirement / Question / Advise by The | | The sixth paragraph A rule system combines the structural components incorporating the confidence supporting sets as the input variables (see page 3, paragraph 2 where the confidence arises by defining the input boundaries) | | |
| Action # | | | | |

| Action # | Requirement / Question / Advise by The Examiner | Analysis & Response from the Applicant |
|----------|--|--|
| | | 'this threshold is based on the forecast, plus a margin of error', then to compare with the daily sales. (another example of analyst subjectivity in analysis, which the proposed method does not require) |
| | | Conclusion: the claims are different from the one in the reference. |
| | The seventh paragraph A rule system maps the state history to the | Analysis: In reference U, Goldenberg has to create a simulation to evaluate their method, 'we devised a statistical simulation approach? (page |
| | output, variables (see Figure 5 where state history is determined by the amount of | 3, section of Evaluating the detection system, the first paragraph). And Goldenberg further assumes a medication sales pattern 'is a |
| | antibiotic being purchases). | three-spike linearly increasing pattern', 'steadily over the first 3 days' (this is another example of analyst subjectivity). Goldenberg then 'measures the spike detection ratio (SDR, Goldenberg et al unpublished results)' (see page 3, paragraph 2 in reference U). Figure 5 shows the SDR. |
| | | In Claim 1, this sentence describes a rule system that maps a set of state variables and their transition history into output variables; and the manning is further defined by equation 12 in later section. Here |
| | | the value of an output variable, as defined by equation 15, is the combination of three elements, the likelihood index of abnormality, the trend indicator, and the notential impact index |
| | | Goldenberg's output is simply a ratio of spike detection. Goldenberg does not mention the state variables, no history of a set of state variables, no likelihood index, no trend indicator, no |

| Action # | Requirement/Question/Advise by The Examiner | Analysis & Response from the Applicant |
|----------|--|--|
| | | potential impact index at all. |
| i j | | Conclusion: The claims are different from the one in the reference. |
| 27 | As per Claim2, Goldenberg teaches the system wherein said measurement scheme includes the calculation of monthly (or weekly, or daily, or seasonally) averaged daily sales for the | Analysis: Figure 1 in Reference U, Goldenberg shows the OTC data noisy nature by plotting the raw data and the de-noising data. |
| | categorized OTC medicines as the base line, from the data in the past at the same place, which is one data set (base line) for supporting the rule system (see Figure 1). | In reference U, Goldenberg does not mention the weekly or monthly averaged daily sales. The word 'weekly' are used twice total by Goldenberg in his paper: (1) one is about the data in 'medical and public health sources which are typically collected weekly or even less frequently,' (page 2 last paragraph), (2) the other is 'weekly effect showing higher sales during weekends'. |
| | | No word 'monthly' ever used in Reference U by Goldenberg. Therefore, Goldenberg does not teach the system as described in claim 1 or 2. Conclusion: |
| | | The claim is different from the one in the reference. |
| | As per claim 3, Goldenberg teaches the system wherein said measurement scheme includes the calculation of the deviation of daily sales in the | Analysis: In reference U, Figure 3, Goldenberg compares the sales data with the threshold which 'is in fact three standard deviations', as |
| | current-month from the base line, and it is measured in change of percentage at the same place, which is another data set (the first | Goldenberg writes in Page 3, the second paragraph (above Figure 3), as a typical statistic approach. Goldenberg compares the raw sale data to their forecasts. His |
| | structural component) for supporting the rule | approach is different from the method defined in Claim 3, which |

| Analysis & Response from the Applicant | defined by equation 6. Conclusion: | The claim is different from the one in the reference. | Quotation of Tracking Grocery Data section, paragraph 1 (of reference U): "Tracking Grocery Data Grocery and OTC medication sales have three main advantages for the detection of an outbreak: First, these datasets are typically very large and rich, including information on each purchased item and in many cases include customer information(e.g., address). They are also available on a more frequent scale, such as daily and even hourly basis, and do not include delays in reporting as compared with medical and public health sources which are typically collected weekly or even less frequently, and might contain delays. Second, the outbreak footprint would probably exist in these data earlier than in medical or public health data, because of self treatment that people usually pursue before seeking medical assistance. Third, although grocery and OTC sales do not measure illness directly, we might infer specific symptoms experienced by purchasers at a relatively early stage of the onset of the disease." | Analysis: Goldenberg does not define the public health status by state variables, does not define the dynamic model that governs the stat transition which describes the dynamic process of public health status. | Conclusion: The claim is different from the one in the reference. | Quotation of page 1 paragraph 3 in reference U: "We begin in the next section by providing background and a characterization of an outbreak of a bioagent, focusing on anthrax. Then we describe traditional data collected |
|--|-------------------------------------|---|--|---|--|---|
| Requirement / Question / Advise by The | | | As per claim 7, Goldenberg further teaches the system wherein said dynamic model of the categorized public health status is defined by the system with a set of state variables and state transitions over the time dimension at a specified place, with which state transitions model the change of the categorized public health status (see page 2, Tracking Grocery Data section, paragraph 1 where the state variables and public health are tracked through antibiotic sales). | | | As per claim 10, Goldenberg further teaches the system |
| Action # | | | 31 | | | 32 |

| Analysis & Response from the Applicant | from medical and public health sources and their ability to detect attacks in a timely fashion, before turning to grocery data and the detection system that we developed. We also introduce a method for evaluating the detection system in the absence of a bioagent footprint in the data, and for tuning the system to the input data. We end with some observations on the usefulness of our approach." | Analysis: Goldenberg does not mention state variables, no state transition, no supporting sets for the state transition either. | Goldenberg mentioned 'input data'. However, Goldenberg's 'input data' is the simulated data, created with his assumptions including the time pattern and spike rising pattern. (see page 4, the second paragraph). | In Claim 10, there are no assumptions for input sets. | However, the applicant mended the sentence in Claim 10 to further clarify what is being mapped, please see the response to Action 23: "the structural components are mapped incorporating their confidence levels." | Conclusion: The claim is different from the one in the reference. | onse of (2) : | Fig. 2 in reference U has two diagrams, and they are the results of Wavelet Transform for prediction. Goldenberg compares the sales data with the threshold which 'is in fact three standard deviations', |
|---|--|---|--|---|---|--|--------------------------------------|---|
| Analysis & Respo | from medical and public fashion, before turning talso introduce a method footprint in the data, and observations on the usef | Analysis: Goldenl transition, no supp | Goldenberg menti data' is the simula the time pattern ar paragraph). | In Claim 10, there | However, the application clarify what is bein "the structural com confidence levels." | Conclusion: The claim is diffe | Analysis and response of (2): | Fig. 2 in reference Wavelet Transfordata with the thres |
| Requirement / Question / Advise by The Examiner | (1) wherein said input sets are the supporting sets for the state transition rule systems (see page 1 paragraph 3); | | | | | | (2) it is mapped from the structural | components incorporating the confidence levels (see figure 2 and page 3 paragraph 2 where statistical quality control creates |
| Action # | | | | | | | | |

| Analysis & Response from the Applicant | as Goldenberg writes in Page 3, paragraph 2 (above Figure 3), as a typical statistic approach, not in state pace, no rule system either. | In Claim 10, the approach is defined in state space, and claim 10 further describes the input sets are supporting sets for state transition rule system, and the input sets are function of the confidence level for each structural component. Equation 11 defines the mathematical relation of them. | Conclusion: The claim is different from the one in the reference. | Analysis of (1): Quotation of page 1 paragraph 3 in reference U: "We begin in the next section by providing background and a characterization of an outbreak of a bioagent, focusing on anthrax. Then we describe traditional data collected from medical and public health sources and their ability to detect attacks in a timely fashion, before turning to grocery data and the detection system that we developed. We also introduce a method for evaluating the detection system in the absence of a bioagent footprint in the data, and for tuning the system to the input data. We end with some observations on the usefulness of our approach." | Response: Goldenberg does not mention input sets, not state transition either; not in page 1 paragraph 3, not in his paper at all. Conclusion of (1): The claim is different from the one in the reference. | Analysis of (2) |
|---|--|--|---|---|---|--------------------------------------|
| Requirement / Question / Advise by The Examiner | confidence levels). | | | (1) As per claim 11, Goldenberg further teaches the system wherein said input sets are the supporting sets for the state transition rule systems (see page 1 paragraph 3); | | (2) it is mapped from the structural |
| Action # | | | | 33 | | |

| Action # | Requirement / Question / Advise by The Examiner | Analysis & Response from the Applicant |
|----------|---|--|
| | levels (see figure 3 and page 3 paragraph 2 where statistical quality control creates confidence levels), | with the threshold which 'is in fact three standard deviations', as Goldenberg writes in Page 3, paragraph 2 (above Figure 3), as a typical statistic approach. Response to (2): Goldenberg does not mention the structural components as described in Claim 11, he does not mention how to incorporate their confidence level either. |
| | (3) where the confidence levels are derived from the historical data sets (see page 3 paragraph 2 where the historical data set are the natural variation), (4) and the confidence supporting sets are found from the cumulated distribution functions with the specified confidence levels (see figure 3 and page 3 paragraph 2 where the specified confidence level is the security band). | Analysis of (3) and (4): See response to action 26 for Quotation of paragraph 2 in page 3 of reference U. Analysis: Goldenberg has a totally different approach. He computes the forecasts, compares the raw sale data to the forecasts plus a margin of error. The historical data set are not the natural variation, although the historical data sets (and the structural components) contain the natural variation. Goldenberg tries to use wavelet transform to model the natural variation. Goldenberg's security band is "in fact three standard deviations of the differences above the denoised series" as he wrote. In contrast, in Claim 11, the confidence supporting sets are functions (defined by equation 7, 8 and 9) which are a significant improvement and new invention, eliminating the analyst's subjectivity in choice of thresholds, and assumptions commonly found in his simulated input data. |
| | | Response to (3) and (4): The applicant can not find how Goldenberg teaches the same |

| V-1 | | |
|---|--|---|
| Analysis & Response from the Applicant | Quotation of page 3 paragraph 1 in reference U: "First, we decompose the denoised series into several "resolutions" by using a discrete (redundant) wavelet transform (ref. 12; cf. the continuous version of wavelets in ref. 13). Each resolution describes a different frequency of the series, but, unlike other transforms (e.g., the cosine and Fourier transform), it retains information on the time that each frequency is present. The resulting series for each resolution are more regular, and thus we use a simple autoregressive model (where the sales at time t aretaken to be a weighted average of previous sales) for predictingeach resolution separately. We then add the predictions to create the forecast of the next day sales. Fig. 2 shows the decomposition of the (preprocessed and denoised) series into five resolutions. For each resolution, we use an autoregressive model for forecasting the next point. Finally, we add the forecasts to obtain the next point in the series, i.e., Fig. 2 also includes the combined forecast of the next day (denoised) sales." Analysis: Goldenberg uses wavelet transformations (not state variables) and ones several "resolutions". He does not say what | wavelet function is used, and the 'resolution' is the results of transform in frequency, it has no physical meaning and no public health contents either. In the claims, including claim 12, in contrast to the reference U, state variables and a dynamic model of state transitions are used; Claim 12 describes how the history of the state variables are mapped into the output sets, and each output variable contains 3 elements, with public health meaning: likelihood, trend indicator, and impact indicator, which Goldenberg does not mention any of them (not in Fig 2 either) nor are any equivalent measures offered. |
| Requirement / Question / Advise by The Examiner | As per claim 12, Goldenberg further teaches the system wherein said output sets are a set of vectors (see page 3 paragraph 1 where the vectors are resolutions), each with three values: likelihood, trend indicator, and impact indicator, where the output sets are mapped from the state history at the study place (see figure 2 where the values are related to their normalized counts). | |
| Action # | 34 | |

| Action # | Requirement / Question / Advise by The Examiner | Analysis & Response from the Applicant |
|----------|---|---|
| | | Conclusion: The claim is different from the one in the reference. |
| | As per claim 13, Goldenberg further teaches the system wherein said rule system that governs the state transitions is the system with sets of logical rules, which evaluate both the logical and numerical functions to determine the system states | Quotation of page 2 paragraph 10 in reference U: "The third layer of the system forecasts the next day sales given all of the previous sales. Although the data are now denoised, simple time-series models (e.g., autoregressive moving average models) do not perform well because of the non-stationarity of the series, i.e., the changes in their behavior over time cannot be characterized by simple time-series models. Instead, we use a two-stage prediction method suitable for non-stationary data that can be easily automated and yields more accurate predictions." |
| | | Analysis: Goldenberg does not say what his two-stage prediction is; he does not mention the state variables, not state transitions either in his paper. And no rule system governs the state transition in his paper. |
| | | Therefore, the claim is different from the one in the reference. |
| | As per claim 14, Goldenberg further teaches the system wherein said rule system that processes the structural components is a rule system with both logical and numerical functions manning the structural | Analysis: Goldenberg uses wavelet transform, not the state variables as in the claim 1-15. In the proposed patent, the rule system processes state variables, with both logical and numerical functions. |
| | components to supporting sets (see figure 2). | Figure 2 shows his wavelet transform results. There is no word of 'rule' or word of 'logic' is used in reference U. |
| | | Therefore, the claim is different from the one in the reference. |
| | As per claim 15, Goldenberg further | Analysis: Goldenberg uses wavelet transform and auto-regression |

| Action # | Requirement // Question / Advise by, The Examiner | Analysis & Response from the Applicant |
|----------|--|---|
| | teaches the system wherein said rule system that maps the state history to the output variables is a rule system with both logical and numerical functions mapping the state variables to the output variables which are described in Claim 12 (see figure 2 where the output variables are the prediction). | to forecast the sales; while the claims use state variable and dynamic model of state transitions, a rule system maps the state history into the output variables, each output variable has 3 elements as describes in Claim 12. Goldenberg does not use state variables, no rule system, nor any of those three elements. See response to Action 34. Therefore, the claim is different from the one in the reference. |
| 38 | Quotation of 35. U.S.C. 103(a) | Noted. |
| 39 | Claim 8 is rejected because of reference U and reference V on the attached. | See response to action 40. |
| 40 | (1) As per claim 8, Goldenberg teaches the apparatus of claim 7 Armstrong teaches the apparatus, (see page 7 Measure paragraph 2) | See response to action 31. Goldenberg does not define the public health status by state variables, does not define the dynamic model that governs the state transitions which describes the dynamic process of public health status. No word of 'state' or 'dynamic model' is used in Reference U. Thus, the applicant can not agree 'Goldenberg teaches the apparatus of claim 7'. |
| 40 | (2) Armstrong teaches the apparatus, (see page 7 Measure paragraph 2) | Quotation of page 7 Measure paragraph 2 in reference V by Armstrong et al: "Measures. Parameters for measuring the importance of a health-related event—and therefore the public health surveillance system with which it is monitored—can include (7) • indices of frequency (e.g., the total number of cases and/or deaths; incidence rates, prevalence, and/or mortality rates); and summary measures of population health status (e.g., quality-adjusted life years [QALYS]); • indices of severity (e.g., bed-disability days, case-fatality ratio, and hospitalization rates and/or disability rates); |

| Analysis & Response from the Applicant | disparities or inequities associated with the health-related event; costs associated with the health-related event; preventability (10); potential clinical course in the absence of an intervention (e.g., vaccinations) (11,12); and public interest. | Analysis: The content described in the application Claim 1-15 is 'a dynamic process of categorized public health status in a community (such as flu, or gastrointestinal diseases) at a place in a specified time, the process changes in hourly and daily, it is totally different from the above reference which measure by years, disability, and cost etc. | Response: Applicant can not see how reference V defined the dynamic process of categorized public health status in a community with set of state variables, as healthy status, critical status, starting-unusual status, upward-trend-unusual status, peak-unusual status, downward-trend status, and ending-unusual status. | The cited reference is delineating the administrative approach to public health status, not introducing a model by which the health status of a population can be described automatically and consistently. The state variables used in the proposed method and the health status indicators used in the cited sources are of a very different nature. One key difference is that the cited indicators do not relate to the field of early detection of public health events. | Quotation of reference V, page 3, Summary paragraph 1 |
|---|---|--|--|---|---|
| Requirement / Question / Advise by The Examiner | | | | | (3) Armstrong page 3, Summary paragraph |
| Action # | | | | | 40 |

| Analysis & Response from the Applicant | Summary | "The purpose of evaluating public health surveillance systems is to ensure that problems of public health importance are being monitored efficiently and effectively. CDC's Guidelines for Evaluating Surveillance Systems are being updated to address the need for a) the integration of surveillance and health information systems, b) the satablishment of data standards, c) the electronic exchange of health information systems, b) the objectives of public health surveillance to facilitate the response of public health of emerging health threats (e.g., new diseases). This report provides updated guidelines for evaluating surveillance systems based on CDC's Framework for Program Evaluation in Public Health, research and discussion of concerns related to public health surveillance systems, and comments received from the public health community. The guidelines in this report describe many tasks and related activities that can be applied to public health surveillance systems." | Response: The applicant fully supports the above Summary, and systematically developed a new method (as detailed in claims 1-15). This method has been implemented in computer systems to improve the public health surveillance in real world. While the success of that implementation system can be assessed through the guidelines proposed in the cited source, the cited source does not disclose or cover the methods introduced by the application in any manner. |
|---|---------|--|--|
| # Requirement / Question / Advise by The :Examiner* | | | |
| Action # | | ·/ | |



UNITAR SEATES PATENT AND TRADEMARK OFFICE

UNITED STATES DEPARTMENT OF COMMERCE United States Patent and Trademark Office Address: COMMISSIONER FOR PATENTS P.O. Box 1450 Alexandria, Virginia 22313-1450 www.uspto.gov

APPLICATION NO. FIRST NAMED INVENTOR ATTORNEY DOCKET NO. CONFIRMATION NO. 10/662,552 09/15/2003 Xiaohui Zhang 2279 12/19/2006 36293 **EXAMINER** SCIENTIFIC TECHNOLOGIES CORPORATION 4400 E BROADWAY BLVD SEREBOFF, NEAL **SUITE 705** ART UNIT PAPER NUMBER TUCSON, AZ 85711 3626

SHORTENED STATUTORY PERIOD OF RESPONSE MAIL DATE DELIVERY MODE

3 MONTHS 12/19/2006 PAPER

Please find below and/or attached an Office communication concerning this application or proceeding.

If NO period for reply is specified above, the maximum statutory period will apply and will expire 6 MONTHS from the mailing date of this communication.